

# **Foundation of Cryptography (0368-4162-01), Lecture 3**

## **Hardcore Predicates for Any One-way Function**

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### Definition 1 (hardcore predicates)

An efficiently computable function  $b : \{0, 1\}^n \mapsto \{0, 1\}$  is an hardcore predicate of  $f : \{0, 1\}^n \mapsto \{0, 1\}^n$ , if

$$\Pr[P(f(U_n)) = b(U_n)] \leq \frac{1}{2} + \text{neg}(n),$$

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## Theorem 2

Let  $f : \{0, 1\}^n \mapsto \{0, 1\}^n$  be a OWF, and define  $g : \{0, 1\}^n \times \{0, 1\}^n \mapsto \{0, 1\}^n \times \{0, 1\}^n$  as  $g(x, r) = f(x), r$ . Then  $b(x, r) = \langle x, r \rangle_2$ , is an hardcore predicate of  $g$ .

Note that if  $f$  is one-to-one, then so is  $g$ .

## Section 1

# The Information Theoretic Case

**Definition 3 (min-entropy)**

The min entropy of a random variable  $X$ , is defined

$$H_{\infty}(X) := \min_{y \in \text{Supp}(X)} \log \frac{1}{\Pr_X[y]}.$$

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Examples

## Pairwise independent hashing

### Definition 4 (pairwise independent hash functions)

A function family  $\mathcal{H}$  from  $\{0, 1\}^n$  to  $\{0, 1\}^m$  is pairwise independent, if for every  $x \neq x' \in \{0, 1\}^n$  and  $y, y' \in \{0, 1\}^m$ , it holds that  $\Pr_{h \leftarrow \mathcal{H}}[h(x) = y \wedge h(x') = y'] = 2^{-2m}$ .

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### Lemma 5 (leftover hash lemma)

Let  $X$  be a random variable over  $\{0, 1\}^n$  with  $H_\infty(X) \geq k$  and let  $\mathcal{H}$  be a family of pairwise independent hash functions from  $\{0, 1\}^n$  to  $\{0, 1\}^m$ , then

$$\text{SD}((h, h(x))_{h \leftarrow \mathcal{H}, x \leftarrow X}, (h, y)_{h \leftarrow \mathcal{H}, y \leftarrow \{0, 1\}^m}) \leq 2^{(m-k-2)/2}.$$



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\* We typically simply write  $\text{SD}((H, H(X)), (H, U_m))$ , where  $H$  is uniformly distributed over  $\mathcal{H}$ .

## efficient function families

### Definition 6 (efficient function family)

An ensemble of function families  $\mathcal{F} = \{\mathcal{F}_n\}_{n \in \mathbb{N}}$  is efficient, if the following hold:

**Samplable.**  $\mathcal{F}$  is samplable in polynomial-time: there exists a PPT that given  $1^n$ , outputs (the description of) a uniform element in  $\mathcal{F}_n$ .

**Efficient.** There exists a polynomial-time algorithm that given  $x \in \{0, 1\}^n$  and (a description of)  $f \in \mathcal{F}_n$ , outputs  $f(x)$ .

## hardcore predicate for regular OWF

### Lemma 7

*Let  $f : \{0, 1\}^n \mapsto \{0, 1\}^n$  be a  $d(n) \in 2^{\omega(\log n)}$  regular function and let  $\mathcal{H} = \{\mathcal{H}_n\}$  be an efficient family of Boolean pairwise independent hash functions over  $\{0, 1\}^n$ . Define*

*$g : \{0, 1\}^n \times \mathcal{H}_n \mapsto \{0, 1\}^n \times \mathcal{H}_n$  as*

$$g(x, h) = (f(x), h),$$

*then  $b(x, h) = h(x)$  is an hardcore predicate of  $g$ .*

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Proof: We prove the claim by showing that

### Claim 8

$SD((f(U_n), H, H(U_n)), (f(U_n), H, U_1)) = \text{neg}(n)$ , where the rv  $H = H(n)$  is uniformly distributed over  $\mathcal{H}_n$ .

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Does this conclude the proof?

## Proving Claim 8

Proof: For  $y \in \{f(x) : x \in \{0, 1\}^n\}$ , let the rv  $X_y$  be uniformly distributed over  $f^{-1}(y) := \{x \in \{0, 1\}^n : f(x) = y\}$ .

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$$\begin{aligned} & \text{SD}((f(U_n), H, H(U_n)), (f(U_n), H, U_1)) \\ &= \sum_{y \in f(\{0, 1\}^n)} \Pr[f(U_n) = y] \cdot \text{SD}((f(U_n), H, H(U_n) \mid f(U_n) = y) \\ & \quad , (f(U_n), H, U_1 \mid f(U_n) = y)) \end{aligned}$$



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## Proving Claim 8 cont.

Since  $H_\infty(X_y) = \log(d(n))$  for any  $y \in \{f(x) : x \in \{0, 1\}^n\}$ ,

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The leftover hash lemma yields that

$$\begin{aligned} \text{SD}((y, H, H(X_y)), (y, H, U_1)) &\leq 2^{(1-H_\infty(X_y)-2)/2} \\ &= 2^{-(\log d(n)+1)/2} = \text{neg}(n). \quad \square \end{aligned}$$

## Further remarks

### Remark 9

- We can output  $\Theta(\log d(n))$  bits,
- $g$  and  $b$  are not defined over all input length.